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# OPT++

## A Toolkit for Nonlinear Optimization

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# Acknowledgements

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- ❖ OPT++ is an open source toolkit for general nonlinear optimization problems
- ❖ Original development started in 1992 at SNL/CA
- ❖ Major contributors
  - Juan Meza, LBNL
  - Patty Hough, SNL/CA
  - Pam Williams, SNL/CA
- ❖ Other members of the OPT++ team
  - Vicki Howle
  - Kevin Long
  - Suzanne Shontz

- ❖ Introduction to Optimization
- ❖ OPT++ Philosophy
- ❖ Classes of Optimization Solvers
- ❖ Setting up a Problem and Algorithm
- ❖ Example 1: Unconstrained Optimization
- ❖ Example 2: Constrained Optimization
- ❖ Parallel optimization techniques
- ❖ Summary

# General Optimization Problem

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$$\min_{x \in \mathcal{R}^n} f(x),$$

Objective function

$$s.t. \quad h(x) = 0,$$

Equality constraints

$$g(x) \geq 0$$

Inequality constraints

$$L = f(x) + y^T h(x) - w^T g(x)$$

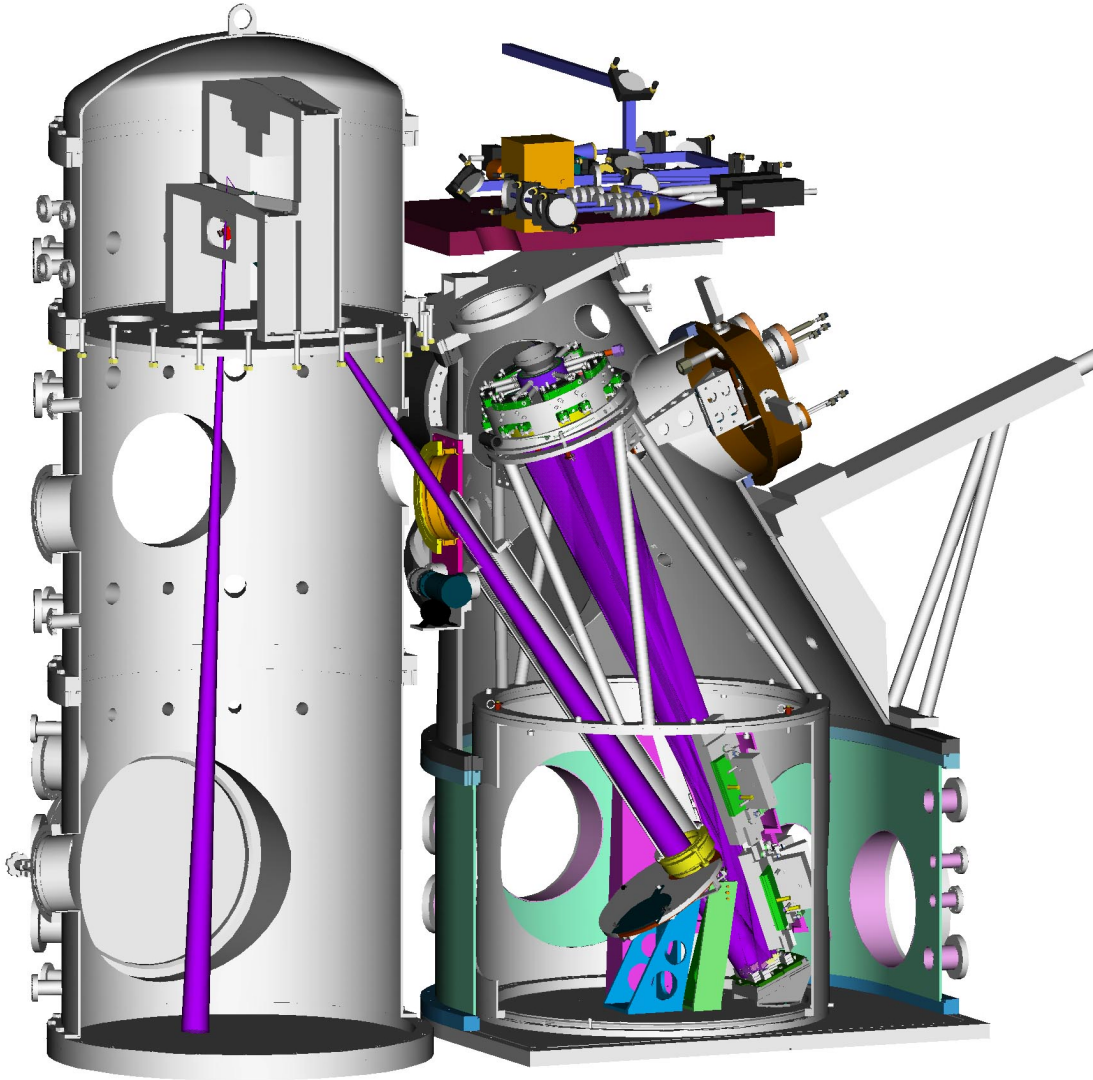
# Classes of Optimization Problems

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- ❖ Unconstrained optimization
- ❖ Bound constrained optimization
  - Only upper and lower bounds
  - Sometimes called “box” constraints
- ❖ General nonlinearly constrained optimization
  - Equality and inequality constraints
  - Usually nonlinear
- ❖ Some special case classes
  - Linear programming (function and constraints linear)
  - Quadratic programming (quadratic function, linear constraints)

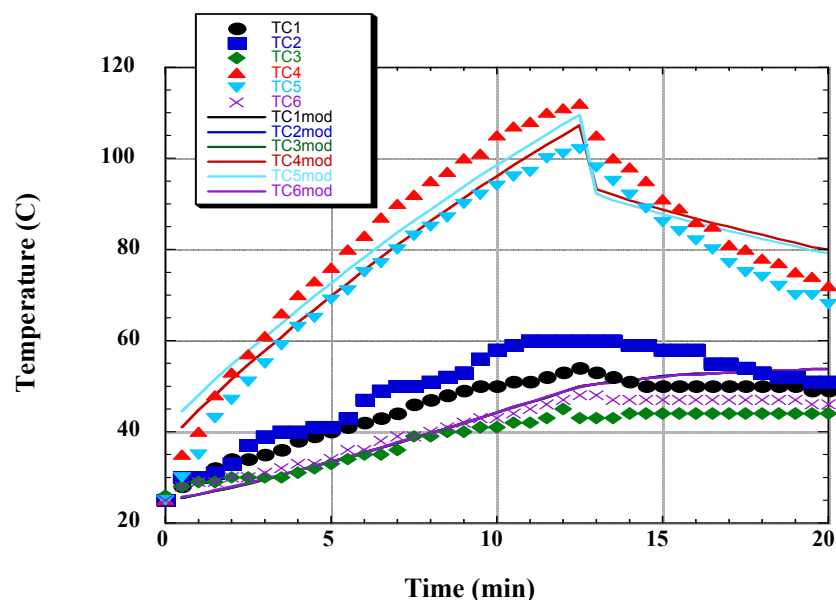
# Parameter identification example



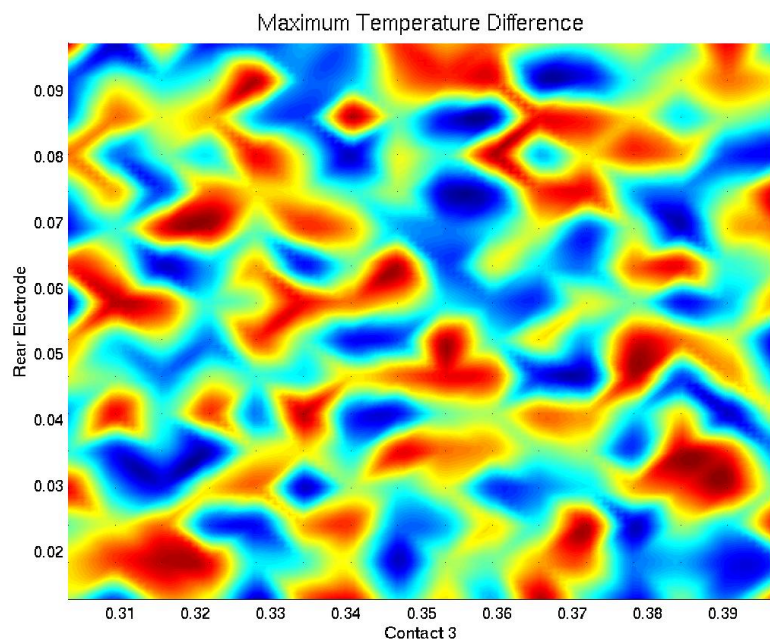
$$\begin{aligned} \min_x \quad & \sum_{i=1}^N (T_i(x) - T_i^*)^2 \\ \text{s. t.} \quad & 0 \leq x \leq u \end{aligned}$$

- ❖ Find model parameters, satisfying some bounds, for which the simulation matches the observed temperature profiles
- ❖ Computing objective function requires running thermal analysis code

# Optimization formulation



- ❖ Objective function consists of computing the max temperature difference over 5 curves
- ❖ Each simulation requires approximately 7 hours on 1 processor
- ❖ Uncertainty in both the measurements and the model parameters



# Some working assumptions

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- ❖ Objective function is smooth
  - Usually true, but simulations can create noisy behavior
- ❖ Twice continuously differentiable
  - Usually true, but difficult to prove
- ❖ Constraints are linearly independent
  - Users can sometimes overspecify or incorrectly guess constraints
- ❖ Small dimensional, but expensive objective functions



- ❖ Problem should be defined in terms the user understands
  - Do I have second derivatives available? vs. Is my objective function twice continuously differentiable?
- ❖ Solution methods should be easily interchangeable
  - Once the problem is setup, methods should be easy to interchange so that the user can compare algorithms
- ❖ Common components of methods should be interchangeable
  - Algorithm developers should be able to re-use common components from other algorithms

## ❖ Four major classes of problems available

- *NLF0(ndim, fcn, init\_fcn, constraint)*
  - Basic nonlinear function, no derivative information available
- *NLF1(ndim, fcn, init\_fcn, constraint)*
  - Nonlinear function, first derivative information available
- *FDNLF1(ndim, fcn, init\_fcn, constraint)*
  - Nonlinear function, first derivative information approximated
- *NLF2(ndim, fcn, init\_fcn, constraint)*
  - Nonlinear function, first and second derivative information available

# Classes of Solvers in OPT++

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## ❖ Pattern search

- No derivative information required

## ❖ Conjugate Gradient

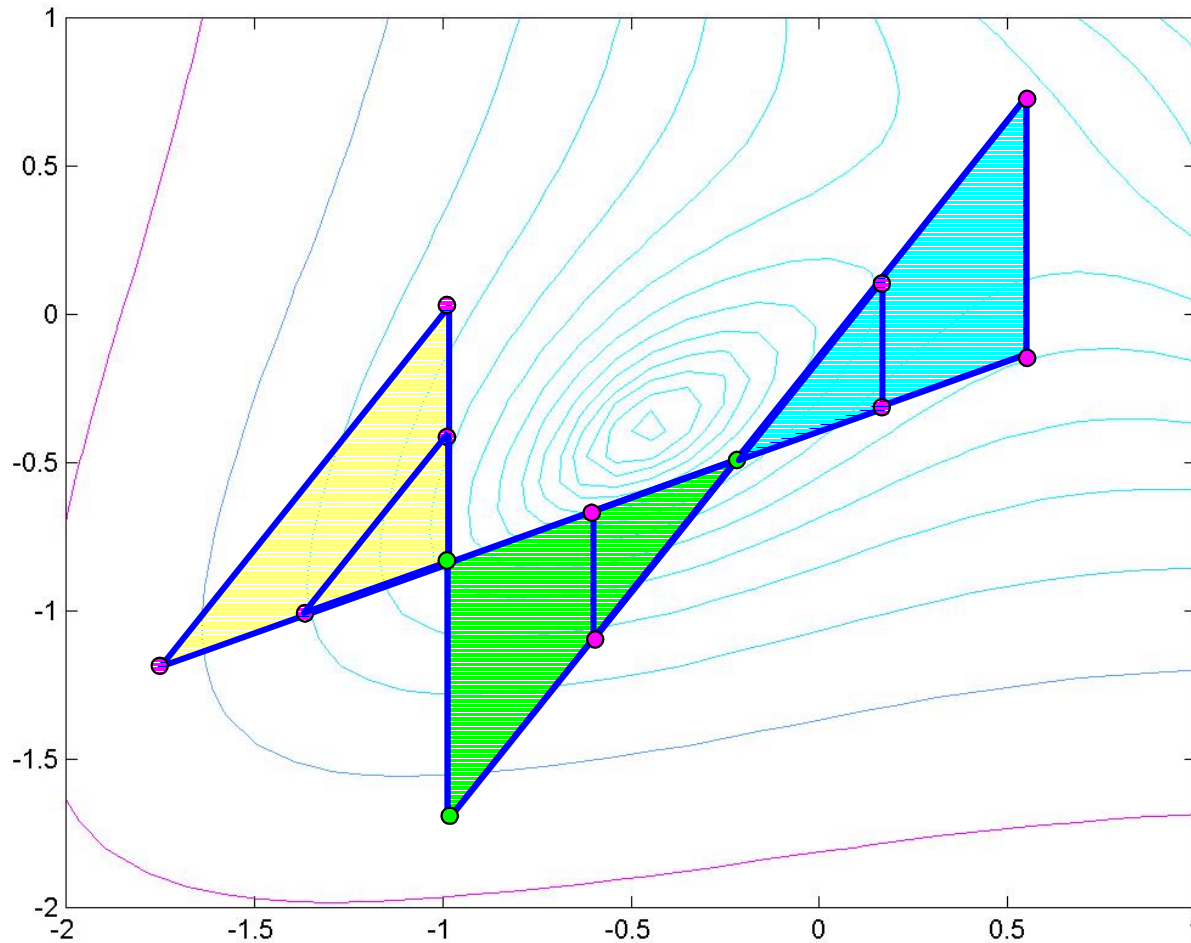
- Derivative information may be available but doesn't use quadratic information

## ❖ Newton-type methods

- Algorithm attempts to use/approximate quadratic information
- Newton
- Finite-Difference Newton
- Quasi-Newton
- NIPS

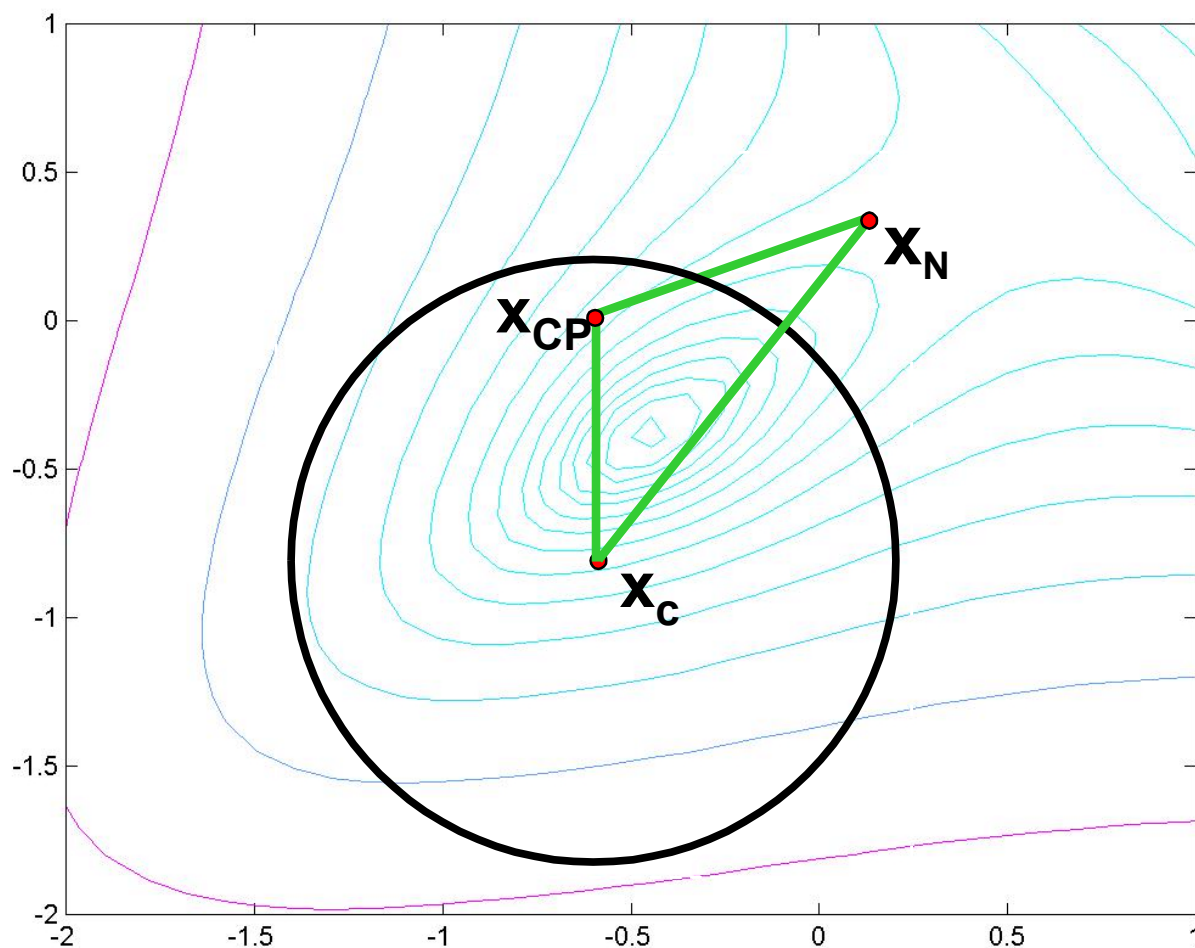
# Quick tour of some of the algorithms

# Pattern search



- ❖ Can handle noisy functions
- ❖ Do not require derivative information
- ❖ Inherently parallel
- ❖ Convergence can be painfully slow

# Newton-type Methods



- ❖ Fast convergence properties
- ❖ Good global convergence properties
- ❖ Inherently serial
- ❖ Difficulties with noisy functions

- ❖ Interior point method
- ❖ Based on Newton's method for a particular system of equations (perturbed KKT equations, slack variable form)
- ❖ Can handle general nonlinear constraints
- ❖ Can handle strict feasibility

$$F(\mu) = \begin{bmatrix} \nabla f(x) + \nabla h(x)y - \nabla g(x)w \\ w - z \\ h(x) \\ g(x) - s \\ ZSe - \mu e \end{bmatrix} = 0$$

## ❖ Constraint types

- `BoundConstraint(numconstraints, lower, upper)`
- `LinearInequality(A, rhs, stdFlag)`
- `NonLinearInequality(nlprob, rhs, numconstraints, stdFlag)`
- `LinearEquation(A, rhs)`
- `NonLinearEquation(nlprob, rhs, numconstraints)`

## ❖ The whole shebang

- `CompoundConstraint(constraints)`



# Algorithm Choices Depend on Problem



	NLF0	FDNLF1	NLF1	NLF2
OptPDS	X	X	X	X
OptCG		X	X	X
OptQNewton		X	X	X
OptBCQNewton		X	X	X
OptFDNewton		X	X	X
OptFDNIPS		X	X	X
OptNewton				X
OptBCNewton				X
OptNIPS				X

# Bare bones example: unconstrained optimization

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```
void init_rosen(int ndim, ColumnVector& x);  
void rosen(int ndim, const ColumnVector& x, double& fx, int& result);  
  
int main() {  
    int ndim = 2;  
    FDNLF1 nlp(ndim, rosen, init_rosen);  
    nlp.initFcn();  
    OptQNewton objfcn(&nlp);  
    objfcn.setSearchStrategy(TrustRegion);  
    objfcn.setMaxFeval(200);  
    objfcn.setFcnTol(1.e-4);  
    objfcn.optimize();  
}
```

# Example 2: Constrained optimization



$$\min (x_1 - x_2)^2 + (1/9)(x_1 + x_2 - 10)^2 + (x_3 - 5)^2$$

*s.t.*

$$x_1^2 + x_2^2 + x_3^2 \leq 48,$$

$$-4.5 \leq x_1 \leq 4.5,$$

$$-4.5 \leq x_2 \leq 4.5,$$

$$-5.0 \leq x_3 \leq 5.0$$

# Constrained optimization (cont.)



```
int ndim = 3;
ColumnVector lower(ndim), upper(ndim);
lower << -4.5 << -4.5 << -5.0;   upper << 4.5 << 4.5 << 5.0 ;
Constraint c1 = new BoundConstraint(ndim, lower, upper);
// Nonlinear inequality constraint
NLP* chs65 = new NLP(new NLF2(ndim, 1, ineq, hs65,
    init_hs65));
Constraint nleqn = new NonLinearInequality(chs65);
// Put everything together in one constraint object
CompoundConstraint* constraints = new
    compoundConstraint(nleqn, c1);
```

# Constrained optimization (cont.)

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// Put it all together

```
NLF2 nips(ndim, hs65, init_hs65, constraints);
```

```
nips.initFcn();
```

// Define the optimization object

```
OptNIPS objfcn(&nips);
```

// Set tolerances and parameters

```
objfcn.setFcnTol(1.0e-06);
```

```
objfcn.setMaxIter(150);
```

```
objfcn.setMeritFcn(ArgaezTapia);
```

```
objfcn.optimize();
```

# Parallel Optimization

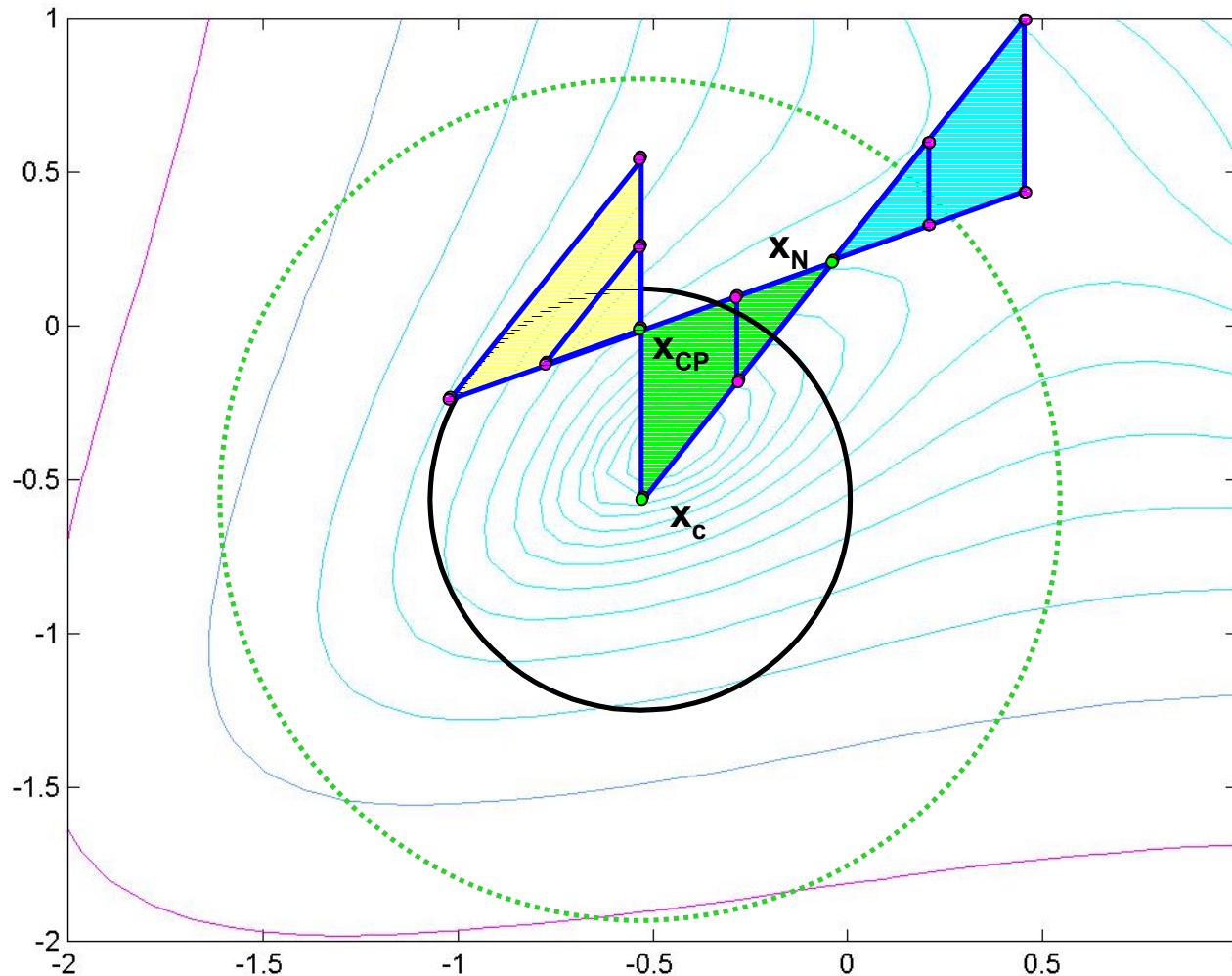
# Schnabel (1995) Identified Three Levels for Introducing Parallelism Into Optimization

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- ❖ Parallelize evaluation of function/gradient/constraints
  - May or may not be easy to implement
- ❖ Parallelize linear algebra
  - Really only useful if the optimization problem is large-scale
- ❖ Parallelize optimization algorithm at a high level
  - Multiple function evaluations in parallel

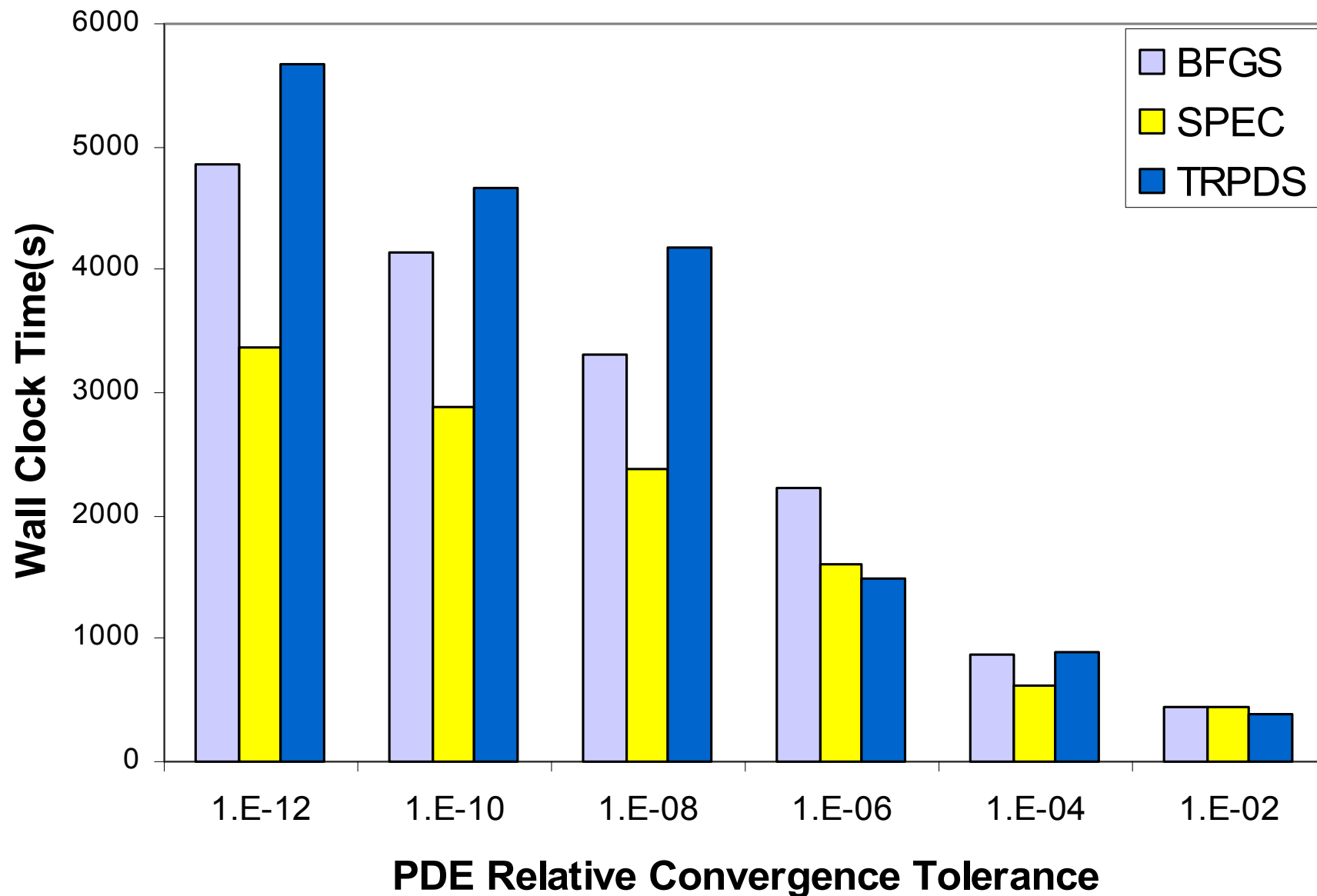
# Trust Region with PDS



- ❖ Fast convergence properties of Newton method
- ❖ Good global convergence properties of trust region approach
- ❖ Inherent parallelism of PDS
- ❖ Ability to handle noisy functions



# Comparison of TRPDS with other approaches



# Summary

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- ❖ OPT++ can handle many types of nonlinear optimization problems
- ❖ The toolkit can be used to compare the effectiveness of several algorithms on the same problem easily
- ❖ The user needs to provide only functions for the objective function and the constraints
  - If additional information is available it can be easily incorporated
- ❖ The code is open source and available at either
  - <http://www.nersc.gov/~meza/projects/opt++>
  - <http://csmr.ca.sandia.gov/opt++>

## ❖ Other links

- <http://sal.kachinatech.com/B/3/index.shtml>
- <http://www-neos.mcs.anl.gov/neos>
- <http://www.mcs.anl.gov/tao>
- <http://endo.sandia.gov/DAKOTA/index.html>

## ❖ Books/Papers

- Dennis and Schnabel, *Numerical Methods for Unconstrained Optimization and Nonlinear Equations*, Prentice-Hall, 1983
- Gill, Murray, Wright, *Practical Optimization*, Academic Press, 1981
- El-Bakry, Tapia, Tsuchiya, Zhang, *On the Formulation and Theory of the Newton Interior-Point Method for Nonlinear Programming*, JOTA, Vol. 89, No.3, pp.507-541, 1996
- More' and Wright, *Optimization Software Guide*, SIAM, 1993